## Homework Set #5

MAC with common information. Consider a DM-MAC P<sub>Y|X1,X2</sub> with three independent uniformly distributed messages W<sub>0</sub> ∈ [1,...,2<sup>nR0</sup>], W<sub>1</sub> ∈ [1,...,2<sup>nR1</sup>], and W<sub>2</sub> ∈ [1,...,2<sup>nR2</sup>]. The first encoder maps each pair (w<sub>0</sub>, w<sub>1</sub>) into a codeword x<sup>n</sup><sub>1</sub>(w<sub>0</sub>, w<sub>1</sub>) and the second maps each pair (w<sub>0</sub>, w<sub>2</sub>) into a codeword x<sup>n</sup><sub>1</sub>(w<sub>0</sub>, w<sub>1</sub>) and the second maps each pair (w<sub>0</sub>, w<sub>2</sub>) into a codeword x<sup>n</sup><sub>2</sub>(w<sub>0</sub>, w<sub>2</sub>). The decoder upon receiving y<sup>n</sup>, finds an estimate (ŵ<sub>0</sub>, ŵ<sub>1</sub>, ŵ<sub>2</sub>) of the messages sent. The probability of decoding error is:

$$P_e^{(n)} = \Pr\left((\hat{W}_0, \hat{W}_1, \hat{W}_2) \neq (W_0, W_1, W_2)\right).$$
(1)

Show that the capacity region for this channel is given by the set of rate triples  $(R_0, R_1, R_2)$  such that

$$R_{1} \leq I(X_{1}; Y | X_{2}, U),$$

$$R_{2} \leq I(X_{2}; Y | X_{1}, U),$$

$$R_{1} + R_{2} \leq I(X_{1}, X_{2}; Y | U),$$

$$R_{0} + R_{1} + R_{2} \leq I(X_{1}, X_{2}; Y),$$
(2)

for some  $p(u)p(x_1|u)p(x_2|u)$ . You need to prove achievability and converse. (Hint: In proving the converse you may use the identification  $U_i = W_0$ .)

## MAC with common information solution:

We show that the capacity is given by the set of rate triples  $(R_0, R_1, R_2)$  such that

$$R_{1} \leq I(X_{1}; Y | X_{2}, U),$$

$$R_{2} \leq I(X_{2}; Y | X_{1}, U),$$

$$R_{1} + R_{2} \leq I(X_{1}, X_{2}; Y | U),$$

$$R_{0} + R_{1} + R_{2} \leq I(X_{1}, X_{2}; Y),$$

for some  $p(u)p(x_1|u)p(x_2|u)$ . Note that this set is convex and therefore there is no need for further convexification.

Proof of achievability: Fix  $p(u)p(x_1|u)p(x_2|u)$ . Generate  $2^{nR_0}$  sequences  $u^n(w_0)$  according to  $p(u^n) = \prod_{i=1}^n p(u_i)$ . For each sequence  $u^n$  generater  $2^{nR_1}$  sequences  $x_1^n(w_0, w_1)$ according to  $p(x_1^n|u^n) = \prod_{i=1}^n p(x_{1i}|u_i)$  and  $2^{nR_2}$  sequences  $x_2^n(w_0, w_2)$  according to  $p(x_2^n|u^n) = \prod_{i=1}^n p(x_{2i}|u_i)$ . To transmit  $(w_0, w_1, w_2)$ . the first transmitter sends  $x_1^n(w_0, w_1)$  and the second transmitter sends  $x_2^n(w_0, w_2)$ .

The decoder upon receiving  $y^n$ , looks for the unique  $(\hat{w}_0, \hat{w}_1, \hat{w}_2)$  such that  $\{(u^n(\hat{w}_0), x_1^n(\hat{w}_0\hat{w}_1), x_2^n(\hat{w}_0\hat{w}_2), y^n) \in A_{\epsilon}^{(n)}\}.$ 

Probability or error: Assume (1, 1, 1) is sent and define the events

$$E_{ijk} = \{ (u^n(i), x_1^n(i, j), x_2^n(i, k), y^n) \in A_{\epsilon}^{(n)} \}$$

Then the probability of decoding error is:

$$P_e^{(n)} \le P(E_{111}^c) + \sum_{i \ne 1, j, k} P(E_{ijk}) + \sum_{j \ne 1} P(E_{1j1}) + \sum_{k \ne 1} P(E_{11k}) + \sum_{j, k \ne 1} P(E_{1jk}) + \sum_{j, k \ne 1} P(E_{1jk}) + \sum_{j, k \ne 1} P(E_{1jk}) + \sum_{j \ne 1} P(E_{1jk}) + \sum$$

 $P(E_{111}^c) \rightarrow 0$  by AEP. Now consider the third term

$$\sum_{j \neq 1} P(E_{1j1}) \leq 2^{nR_1} \sum_{\substack{(u^n, x_1^n, x_2^n, y^n) \in A_{\epsilon}^{(n)} \\ \leq}} p(u^n) p(x_1^n | u^n) p(x_2^n | u^n) p(y^n | x_2^n, u^n)$$

where

(a) follows from the jointly typical set, when u and  $x_2$  are known.

Thus if  $R_1 < I(X_1; Y|X_2, U) - 7\epsilon$ , the third term in the bound on  $P_e^{(n)}$  approaches 0 as  $n \to \infty$ . The fourth term follows similarly and we obtain the requirement that  $R_2 \leq I(X_2; Y|X_1, U) - 7\epsilon$ . The last term approaches 0 as  $n \to \infty$  if  $R_1 + R_2 < I(X_1, X_2; Y|U) - 7\epsilon$ . Finally consider the second term

$$\sum_{i \neq 1, j, k} P(E_{ijk}) \leq 2^{n(R_0 + R_1 + R_2)} \sum_{\substack{(u^n, x_1^n, x_2^n, y^n) \in A_{\epsilon}^{(n)} \\ \leq 2^{n(R_0 + R_1 + R_2)} 2^{n(H(U, X_1, X_2, Y) + \epsilon)} 2^{-n(H(U) - \epsilon)} 2^{-n(H(X_1|U) - 2\epsilon)} 2^{-n(H(X_2|U) - 2\epsilon)} 2^{-n(H(Y) - \epsilon)}$$

Thus if

$$R_0 + R_1 + R_2 < H(U, X_1, X_2, Y) - H(U) - H(X_1|U) - H(X_2|U) - H(Y) - 7\epsilon$$
  
=  $I(X_1, X_2; Y) - 7\epsilon$ ,

the second sum of the bound  $\mathrm{Pr}_e^{(n)}$  goes to 0 as  $n \to \infty.$ 

Proof of converse:

First consider

$$nR_1 = H(W_1)$$
  
=  $H(W_1|W_0, W_2)$   
 $\leq I(W_1; Y^n|W_0, W_2) + n\epsilon_n$ 

$$= \sum_{i=1}^{n} I(W_{1}; Y_{i}|W_{0}, W_{2}, Y^{i-1}) + n\epsilon_{n}$$

$$\leq \sum_{i=1}^{n} H(Y_{i}|W_{0}, W_{2}) - H(Y_{i}|W_{0}, W_{1}, W_{2}, Y^{i-1}) + n\epsilon_{n}$$

$$\leq \sum_{i=1}^{n} H(Y_{i}|X_{2i}, W_{0}, W_{2}) - H(Y_{i}|X_{1i}, X_{2i}, W_{0}, W_{1}, W_{2}, Y^{i-1}) + n\epsilon_{n}$$

$$\leq \sum_{i=1}^{n} H(Y_{i}|W_{0}, X_{2i}) - H(Y_{i}|W_{0}, X_{1i}, X_{2i}) + n\epsilon_{n}$$

$$= \sum_{i=1}^{n} I(X_{1i}; Y_{i}|X_{2i}, U_{i}) + n\epsilon_{n},$$

where  $U_i = W_0$ , so  $p(u_i, x_{1i}, x_{2i}) = p(u_i)p(x_{1i}|u_i)p(x_{2i}|u_i)$ . Similarly it can be shown that  $nR_2 \leq \sum_{i=1}^n I(X_{2i}; Y_i|X_{1i}, U_i) + n\epsilon_n$  and  $n(R_1 + R_2) \leq \sum_{i=1}^n I(X_{1i}, X_{2i}; Y_i|U_i) + n\epsilon_n$ . It is also easy to show that  $n(R_0 + R_1 + R_2) \leq \sum_{i=1}^n I(X_{1i}, X_{2i}; Y_i) + n\epsilon_n$ . Now, introducing a time-sharing random variable Q, we have

$$\begin{array}{rcl} R_{1} & \leq & I(X_{1Q};Y_{Q}|X_{2Q},U_{Q},Q) = I(X_{1};Y|X_{2},U) + \epsilon, \\ R_{1} & \leq & I(X_{2Q};Y_{Q}|X_{1Q},U_{Q},Q) = I(X_{2};Y|X_{1},U) + \epsilon, \\ R_{1} + R_{2} & \leq & I(X_{1Q},X_{2Q};Y_{Q}|U_{Q},Q) = I(X_{1},X_{2};Y|U) + \epsilon, \\ R_{0} + R_{1} + R_{2} & \leq & I(X_{1Q},X_{2Q};Y_{Q}|Q) = I(X_{1},X2;Y) + \epsilon, \end{array}$$

where  $U = (U_Q, Q), X_1 = X_{1Q}, X_2 = X_{2Q}$  and  $Y = Y_Q$ , and

$$p(u, x_1, x_2, y) = p(u)p(x_1|u)p(x_2|u)p(y|x_1, x_2).$$

- 2) Strong ε-typicality. Achievability proofs involving *covering*, e.g., for the rate distortion theorem, require that we find a good lower bound on the probability that one specific typical sequence x<sup>n</sup> is jointly typical with a randomly drawn sequence Y<sup>n</sup>. Using strong typicality, the desired lower bound can be established. Let (X<sub>i</sub>, Y<sub>i</sub>) be drawn i.i.d. ~ P(x, y) and assume that the cardinalities X, Y are finite. Let the marginals of X and Y be P(x) and P(y), respectively (you may use ideas and results from methods of types to solve the exercise). We use in this exercise a specific notation δ(ε) that implies that δ(ε) → 0 as ε → 0.
  - a) Show that if  $x^n \in T_{\epsilon}^{(n)}(X)$ , then

$$p(x^n) \doteq 2^{n(H(X)\pm\delta(\epsilon))}.$$
(3)

- b) Show that  $\Pr(X^n \in T_{\epsilon}^{(n)}(X)) \to 1$ , as  $n \to \infty$ .
- c) Show that

$$|T_{\epsilon}^{(n)}(X)| \doteq 2^{n(H(X)\pm\delta(\epsilon)}.$$
(4)

d) Let  $x^n \in T_{\epsilon}^{(n)}(X)$ , and let  $T_{\epsilon}^{(n)}(Y|x^n)$  be the set of  $y^n$  sequences such that  $(x^n, y^n) \in T_{\epsilon}^{(n)}(X, Y)$ . Show that

$$|T_{\epsilon}^{(n)}(Y|x^n)| \doteq 2^{n(H(Y|X)\pm\delta(\epsilon)}.$$
(5)

e) Let  $x^n \in T^{(n)}_{\epsilon}(X)$ , and  $Y^n$  be drawn independently of  $x^n$  i.i.d.  $\sim P(y)$ . Show that

$$\Pr(x^n, Y^n \in T_{\epsilon}^{(n)}(X, Y) | \doteq 2^{n(I(X,Y) \pm \delta(\epsilon)}.$$
(6)

(Note that in (d) and (e) the bounds do not depend on  $x^n$ .)

## Strong *\epsilon*-typicality solution:

a) We recall that  $T_{\epsilon}^{n}(X) = \{x^{n} : |P_{x^{n}}(a) - p_{X}(a)| < \epsilon, p_{X}(a) = 0 \Rightarrow P_{x^{n}}(a) = 0\}.$ log is a continues function thus  $|x - y| < \epsilon \rightarrow |\log(x) - \log(y)| = |\log \frac{x}{y}| < \delta(\epsilon)$ , and so we obtain-

$$D(P_{X^n}||p(x)) = \sum_{x \in \mathcal{X}} P_{X^n} \log \frac{P_{X^n}}{p(x)}$$

$$\stackrel{(a)}{<} \sum_{x \in \mathcal{X}} P_{X^n} \delta(\epsilon)$$

$$= |\mathcal{X}| \delta(\epsilon) = \delta$$

where

(a) follows from the continuity of the log function, as explained above.

From method of types, we have-  $Q^n(x^n) \doteq 2^{-n(D(p||q)+H(p))}$ , thus for  $x^n \in T^n_{\epsilon}(X)$ , we have-

$$p(x^{n}) = 2^{-n(D(P_{x^{n}}||p(x)) + H(X))} \ge 2^{-n(H(X) + \epsilon)}.$$

It is also clear that-

$$p(x^n) \le 2^{-n(H(X)-\epsilon)},$$

thus we obtain  $p(x^n) \doteq 2^{-n(H(X)\pm\epsilon)}$ .

b) We consider-

$$\begin{split} 1 - \Pr(x^n \in T^n_{\epsilon}(X)) &= & \Pr(x^n \notin T^n_{\epsilon}(X)) \\ &= & \sum_{p(x^n):D(P_{x^n}||p(x)) > \epsilon} p(x^n) \\ &\stackrel{(a)}{=} & \sum_{p(x^n):D(P_{x^n}||p(x)) > \epsilon} 2^{-n(H(X) + \epsilon)} \\ &\leq & |\mathbb{P}_n| 2^{-n(H(X) + \epsilon)} \\ &\stackrel{(b)}{\leq} & (n+1)^{|\mathcal{X}|} 2^{-n(H(X) + \epsilon)} \end{split}$$

$$= 2^{-n(H(X)+\epsilon-\frac{\log(n+1)}{n}|\mathcal{X}|)} \to 0.$$

where

- (a) follows from strong  $\epsilon$  typicality.
- (b) follows from the fact that there are, as an upper bound,  $(n+1)^{|\mathcal{X}|}$  types.

Thus we obtain  $\Pr(x^n \in T^n_{\epsilon}(X)) \to 1$  as  $n \to \infty$ .

c) Using the result from (b) we have-

$$\Pr(x^n \in T^n_{\epsilon}(X)) = \sum_{\substack{x^n \in T^n_{\epsilon}(X) \\ \epsilon}} p(x^n)$$
$$\doteq \sum_{\substack{x^n \in T^n_{\epsilon}(X) \\ \epsilon}} 2^{-n(H(X)\pm\epsilon)}$$
$$= |T^n_{\epsilon}(X)| 2^{-n(H(X)\pm\epsilon)}.$$

Note that if  $\lim a_n = 1 \Rightarrow a_n \doteq 1$ , thus-

$$1 \stackrel{:}{=} \Pr(x^n \in T^n_{\epsilon}(X))$$
$$\stackrel{:}{=} |T^n_{\epsilon}(X)| 2^{-n(H(X)\pm\epsilon)}.$$

Thus we obtain  $|T_{\epsilon}^n(X)| \doteq 2^{-n(H(X)\pm\epsilon)}$ .

d) Let  $x^n \in T^n_{\epsilon}(X)$ , and  $y^n$  is drawn~  $p(y_i|x_i)$ , thus  $(x_i, y_i)$  is drawn i.i.d  $p_{X,Y}$ , and from (a) we have  $Pr((x^n, y^n) \in T^n_{\epsilon}(X, Y)) \to 1$ . Thus-

$$1 \stackrel{:}{=} \Pr((x^n, y^n) \in T^n_{\epsilon}(X, Y))$$

$$\stackrel{(a)}{=} \sum_{x^n \in T^n_{\epsilon}(X)} p((x^n, y^n) \in T^n_{\epsilon}(Y|x^n))$$

$$\stackrel{(b)}{=} \sum_{x^n \in T^n_{\epsilon}(X)} \sum_{y^n \in T^n_{\epsilon}(Y|x^n)} p(x^n, y^n)$$

$$\stackrel{(c)}{=} \sum_{x^n \in T^n_{\epsilon}(X)} \sum_{y^n \in T^n_{\epsilon}(Y|x^n)} 2^{-n(H(X,Y)\pm\epsilon)}$$

$$= \sum_{x^n \in T^n_{\epsilon}(X)} |T^n_{\epsilon}(Y|x^n)| 2^{-n(H(X,Y)\pm\epsilon)}$$

$$= |T^n_{\epsilon}(X)| |T^n_{\epsilon}(Y|x^n)| 2^{-n(H(X,Y)\pm\epsilon)}$$

$$\stackrel{(c)}{=} |T^n_{\epsilon}(Y|x^n)| 2^{-n(H(X,Y)\pm\epsilon)}$$

where

- (a) follows from the definition of  $T^n_{\epsilon}(X,Y)$ .
- (b) follows from the definition of  $T_{\epsilon}^{n}(Y|x^{n})$ .

(c) follows from the fact that  $(x^n, y^n) \in T^n_{\epsilon}(X, Y)$ .

Thus we obtain  $|T_{\epsilon}^n(Y|x^n)| \doteq 2^{n(H(X,Y)-H(X)\pm\epsilon)} = 2^{n(H(Y|X)\pm\epsilon)}.$ 

e) Now we have  $x^n \in T^n_{\epsilon}(X)$ ,  $Y^n \sim p(y)$  i.i.d, independent of  $x^n$ .

$$\begin{aligned} \Pr((x^n, Y^n) \in T^n_{\epsilon}(X, Y)) &\stackrel{(a)}{=} & \sum_{y^n \in T^n_{\epsilon}(Y|x^n)} p(y^n) \\ &\stackrel{(b)}{=} & \sum_{y^n \in T^n_{\epsilon}(Y|x^n)} 2^{-n(H(Y)\pm\epsilon)} \\ &= & |T^n_{\epsilon}(Y|x^n)| 2^{-n(H(Y)\pm\epsilon)} \\ &= & 2^{-n(H(Y)\pm\epsilon-H(Y|X)\pm\epsilon')} \\ &= & 2^{-n(I(X;Y)\pm\epsilon)}. \end{aligned}$$

where

- (a) follows from the definition of  $T_{\epsilon}^{n}(Y|x^{n})$ .
- (b) follows from the fact that  $y^n \in T^n_\epsilon(Y)$ .